

## Exercise 16

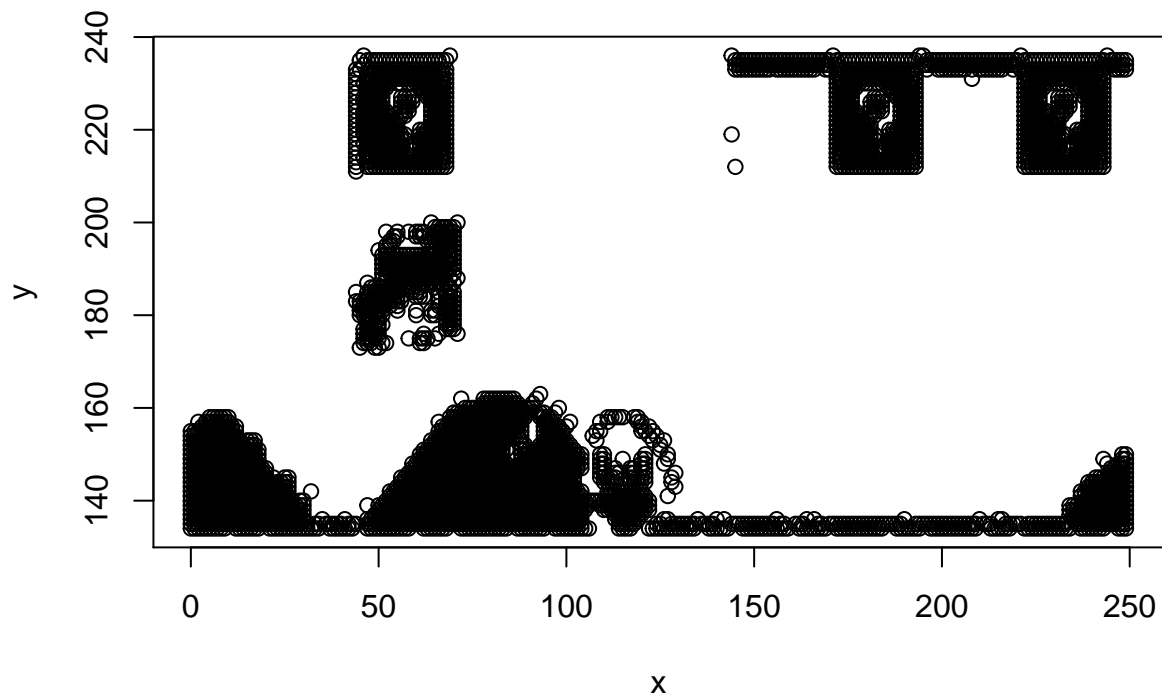
Thip Rattanavilay

2/12/2021

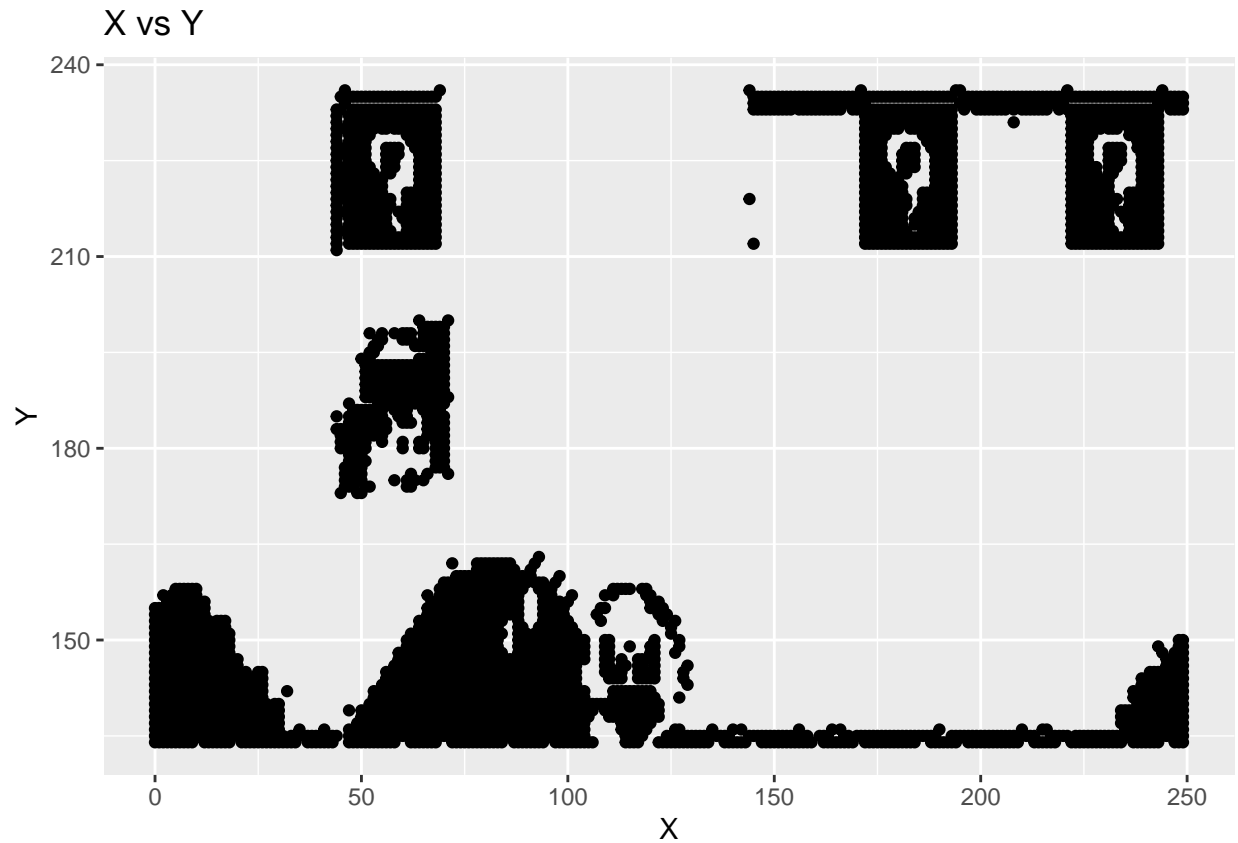
### Clustering

A. Plot the dataset using a scatter plot.

```
plot(datasetcluster_df)
```

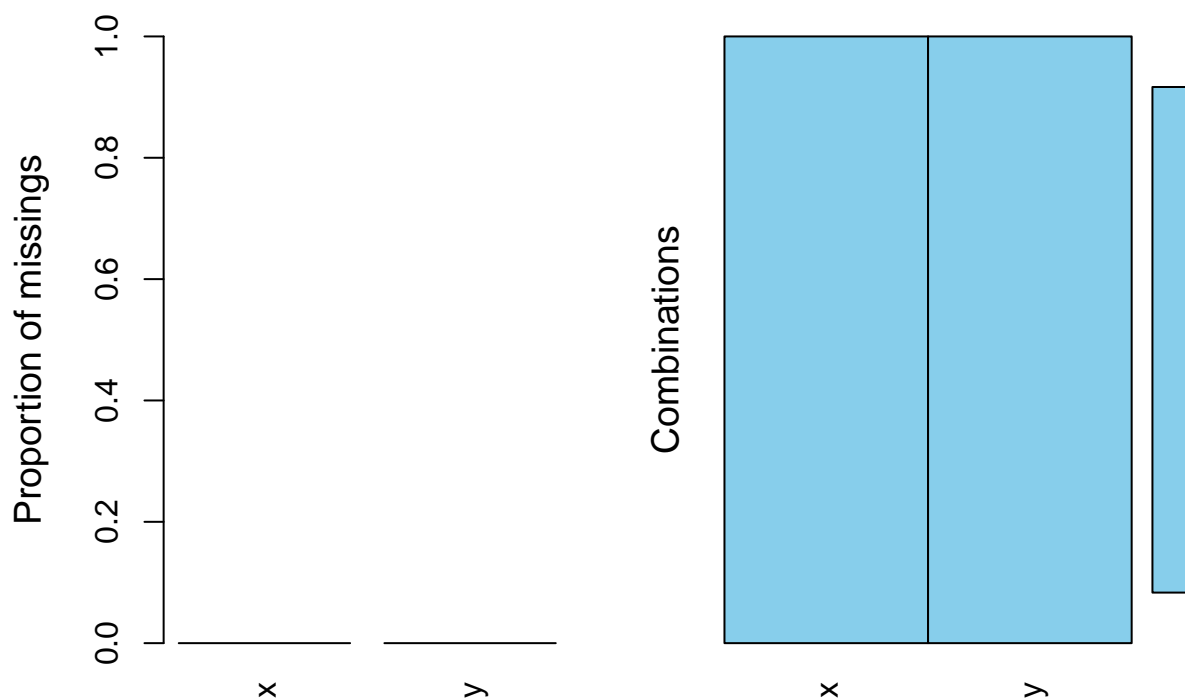


```
ggplot(datasetcluster_df, aes(x= x, y= y)) + geom_point( ) +  
  labs(title= "X vs Y", x="X", y="Y")
```



B. Fit the dataset using the k-means algorithm from  $k=2$  to  $k=12$ . Create a scatter plot of the resultant clusters for each value of  $k$ .

```
aggr(datasetcluster_df)
```



```
cluster2 <- kmeans(datasetcluster_df[, 1:2], 2)
cluster2
```

[illegible]

[illegible]



[illegible]

[illegible]

```

## [3701] 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2
## [3738] 2 2 2 2 2 2 2 2 2 2 2 2 2 2 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
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## [3812] 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3
## [3849] 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 2 2 2 2 2 2 2 2 2
## [3886] 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2
## [3923] 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 1 1 1 1 1 1 1 1
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## [3997] 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
##
## Within cluster sum of squares by cluster:
## [1] 2931827.1 3354777.6 125040.2
## (between_SS / total_SS = 77.6 %)
##
## Available components:
##
## [1] "cluster"      "centers"      "totss"        "withinss"     "tot.withinss"
## [6] "betweenss"    "size"         "iter"         "ifault"

```

```

cluster4 <- kmeans(datasetcluster_df[, 1:2], 4 )
cluster4

```

```

## K-means clustering with 4 clusters of sizes 1272, 1376, 782, 592
##
## Cluster means:
##           x           y
## 1 209.19733 205.2351
## 2  88.12209 144.5305
## 3  57.42711 207.3043
## 4  14.47973 142.9662
##
## Clustering vector:
## [1] 3 3 1 1 1 1 1 1 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3
## [38] 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
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[illegible]



##

[illegible]

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## [3738] 5 5 5 5 5 5 5 5 5 5 5 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3
## [3775] 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3
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## [3849] 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 5 5 5 5 5 5 5 5 5 5 5 5

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```
## [3886] 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5
## [3923] 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 3 3 3 3 3 3 3 3 3 3 3
## [3960] 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3
## [3997] 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3
##
## Within cluster sum of squares by cluster:
## [1] 831375.78 97541.82 267545.20 334024.91 641125.13
## (between_SS / total_SS = 92.4 %)
##
## Available components:
##
## [1] "cluster"      "centers"      "totss"        "withinss"     "tot.withinss"
## [6] "betweenss"    "size"         "iter"         "ifault"
```

[illegible]

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##	[1111]	3	3	3	3	3	3	3	3	3	3	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	3	3
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##	[1629]	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2
##	[1666]	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2
##	[1703]	2	2	2	2																									



[illegible]



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## [1925] 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 1 1 1 1 1 1 4 4 4 4 4 4 4 4 4 4
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## [1999] 4 4 4 4 4 4 4 4 4 4 4 4 4 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 1
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[illegible]



[illegible]

[illegible]

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## [3812] 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3
## [3849] 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5
## [3886] 5 5 5 5 5 5 5 5 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 7 7 7 7 7 7 7 7 7 7
## [3923] 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4
## [3960] 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 1 1 1 1
## [3997] 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
##
## Within cluster sum of squares by cluster:
## [1] 22577.30 35774.69 67391.59 34946.97 61054.03 831375.78 33365.39
## [8] 47345.46 73370.50
## (between_SS / total_SS = 95.8 %)
##
## Available components:
##
## [1] "cluster"      "centers"      "totss"        "withinss"     "tot.withinss"
## [6] "betweenss"    "size"         "iter"         "ifault"       "
```

```
cluster10 <- kmeans(datasetcluster_df[, 1:2],10)
cluster10
```

```
## K-means clustering with 10 clusters of sizes 204, 230, 600, 107, 985, 196, 216, 169, 760, 555
##
## Cluster means:
##           x           y
## 1    8.372549 150.8382
## 2  118.765217 141.4783
## 3   89.816667 146.4567
## 4  179.355140 134.5514
## 5  203.985787 224.8406
## 6    7.806122 139.3878
## 7  239.287037 139.0648
## 8   24.852071 138.5562
## 9   57.410526 208.2447
## 10  66.227027 145.2090
##
## Clustering vector:
## [1] 9 9 5 5 5 5 5 5 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9
## [25] 9 9 9 9 9 9 9 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5
## [49] 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5
## [73] 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5
## [97] 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5
## [121] 5 5 5 5 5 5 5 5 5 5 5 5 5 5 9 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5
## [145] 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5
## [169] 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5
## [193] 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9
## [217] 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5
## [241] 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5
## [265] 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5
## [289] 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 9 9 9 9 9 9 9 9 9 9 9 9 9 9
## [313] 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5
## [337] 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5
## [361] 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9
## [385] 9 9 9 9 9 9 9 9 9 9 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5
```

```

## [409] 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5
## [433] 5 5 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9
## [457] 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5
## [481] 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 9 9 9 9 9 9 9 9
## [505] 9 9 9 9 9 9 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5
## [529] 5 5 5 5 5 5 5 5 5 5 5 9 9 9 9 9 9 9 9 9 9 9 9 5
## [553] 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 9 9
## [577] 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 5 5 5 5 5 5 5 5
## [601] 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 9 9 9 9
## [625] 9 9 9 9 9 9 9 9 9 9 9 9 5 5 5 5 5 5 5 5 5 5 5 5
## [649] 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 9 9 9 9 9 9 9 9
## [673] 9 9 9 9 9 9 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5
## [697] 5 5 5 5 5 5 5 5 5 5 5 9 9 9 9 9 9 9 9 9 9 9 9 9
## [721] 9 9 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5
## [745] 5 5 5 5 5 5 5 5 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9
## [769] 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5
## [793] 5 5 5 5 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 5 5 5 5
## [817] 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5
## [841] 5 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 5 5 5 5 5 5
## [865] 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 9
## [889] 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 5 5 5 5 5 5
## [913] 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5
## [937] 5 5 5 5 5 5 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9
## [961] 9 9 9 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5
## [985] 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 9 9 9 9 9
## [1009] 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 5 5 5 5 5 5 5 5
## [1033] 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5
## [1057] 5 5 5 5 5 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9
## [1081] 9 9 9 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5
## [1105] 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 9 9 9 9 9 9
## [1129] 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 5 5 5 5 5 5 5 5 5
## [1153] 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5
## [1177] 5 5 5 5 5 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9
## [1201] 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5
## [1225] 5 5 5 5 5 5 5 5 5 5 5 5 5 5 9 9 9 9 9 9 9 9 9 9
## [1249] 9 9 9 9 9 9 9 9 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5
## [1273] 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 9 9 9 9 9 9
## [1297] 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 5 5 5 5 5 5 5 5
## [1321] 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5
## [1345] 5 5 5 5 5 5 5 5 5 5 5 5 5 9 9 9 9 9 9 9 9 9 9 9
## [1369] 9 9 9 9 9 9 9 9 9 9 9 9 5 5 5 5 5 5 5 5 5 5 5 5
## [1393] 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5
## [1417] 5 5 5 5 5 5 5 5 5 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9
## [1441] 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9
## [1465] 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9
## [1489] 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9
## [1513] 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9
## [1537] 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9
## [1561] 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9
## [1585] 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9
## [1609] 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9
## [1633] 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9
## [1657] 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9
## [1681] 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9

```



```

## [1705] 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9
## [1729] 9 9 9 9 9 9 9 9 10 10 10 9 9 9 9 9 10 10 10 9 9 9 9 10
## [1753] 10 10 10 10 10 10 10 10 10 10 10 10 10 10 3 10 3 3 3 3 3 3 3
## [1777] 3 3 3 3 3 3 3 3 3 3 3 3 3 3 10 10 10 10 10 3 3 3 3 3
## [1801] 3 3 3 3 3 3 3 3 3 3 3 10 10 10 10 10 10 10 3 3 3 3 3 3
## [1825] 3 3 3 3 3 3 3 3 3 3 3 1 1 1 1 1 1 10 10 10 10 10 10
## [1849] 10 10 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 2 2 2 2 2 2
## [1873] 2 1 1 1 1 1 1 1 1 10 10 10 10 10 10 10 10 10 3 3 3 3 3
## [1897] 3 3 3 3 3 3 3 3 3 3 2 2 2 2 1 1 1 1 1 1 1 1 1 1
## [1921] 10 10 10 10 10 10 10 10 10 10 10 3 3 3 3 3 3 3 3 3 3 3 3
## [1945] 3 3 3 3 2 2 1 1 1 1 1 1 1 1 1 1 1 1 10 10 10 10 10
## [1969] 10 10 10 10 10 10 10 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3
## [1993] 2 2 2 2 2 1 1 1 1 1 1 1 1 1 1 1 1 1 10 10 10 10 10
## [2017] 10 10 10 10 10 10 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 2
## [2041] 2 2 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 10 10 10 10 10
## [2065] 10 10 10 10 10 10 10 10 10 3 3 3 3 3 3 3 3 3 3 3 3 3 3
## [2089] 2 2 2 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 10 10 10
## [2113] 10 10 10 10 10 10 10 10 10 10 10 3 3 3 3 3 3 3 3 3 3 3 3
## [2137] 3 3 3 3 2 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
## [2161] 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 3 3 3 3 3 3 3
## [2185] 3 3 3 3 3 3 3 3 3 3 3 3 3 2 1 1 1 1 1 1 1 1 1
## [2209] 1 1 1 1 1 1 1 1 1 1 10 10 10 10 10 10 10 10 10 10 10 10
## [2233] 10 10 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3
## [2257] 3 2 2 2 2 7 7 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
## [2281] 1 1 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 3 3 3 3 3
## [2305] 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 2 2 2 2 2
## [2329] 7 7 7 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 10 10
## [2353] 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 3 3 3 3 3 3 3
## [2377] 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 2 2 2 2 2 7 7
## [2401] 7 7 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 8 8 10
## [2425] 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 3 3 3 3 3 3 3
## [2449] 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 2 2 2
## [2473] 2 2 2 2 7 7 7 7 7 1 1 1 1 1 1 1 1 1 1 1 1 1 1
## [2497] 1 1 1 8 8 8 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10
## [2521] 10 10 10 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3
## [2545] 3 3 3 2 2 2 2 2 2 2 2 7 7 7 7 7 6 6 6 6 6 6 6
## [2569] 6 6 6 1 1 1 1 1 1 8 8 8 8 8 8 8 10 10 10 10 10 10 10
## [2593] 10 10 10 10 10 10 10 10 10 10 10 10 10 10 3 3 3 3 3 3 3 3
## [2617] 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 2 2 2 2 2 2 2 2
## [2641] 2 2 7 7 7 7 7 7 7 7 7 6 6 6 6 6 6 6 6 6 6 6 6
## [2665] 6 6 6 6 8 8 8 8 8 8 8 8 8 10 10 10 10 10 10 10 10 10
## [2689] 10 10 10 10 10 10 10 10 10 10 10 10 3 3 3 3 3 3 3 3 3 3 3
## [2713] 3 3 3 3 3 3 3 3 3 3 3 3 3 2 2 2 2 2 2 2 2 2 7 7
## [2737] 7 7 7 7 7 7 7 7 7 6 6 6 6 6 6 6 6 6 6 6 6 6 6
## [2761] 6 6 8 8 8 8 8 8 8 8 8 8 10 10 10 10 10 10 10 10 10 10
## [2785] 10 10 10 10 10 10 10 10 10 10 10 3 3 3 3 3 3 3 3 3 3 3 3
## [2809] 3 3 3 3 3 3 3 3 3 3 3 3 2 7 7 7 7 7 7 7 7 7 6
## [2833] 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 8 8 8 8 8 8 8
## [2857] 8 8 8 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10
## [2881] 10 10 10 10 10 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3
## [2905] 3 3 3 3 3 3 2 2 2 2 2 2 2 2 2 2 7 7 7 7 7 7 7
## [2929] 7 7 7 7 7 7 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 8
## [2953] 8 8 8 8 8 8 8 8 8 10 10 10 10 10 10 10 10 10 10 10 10 10
## [2977] 10 10 10 10 10 10 10 10 10 10 10 3 3 3 3 3 3 3 3 3 3 3 3

```

```

## [3001] 3 3 3 3 3 3 3 3 3 3 3 3 2 2 2 2 2 2 2 2 2 2
## [3025] 2 7 7 7 7 7 7 7 7 7 7 7 7 7 7 6 6 6 6 6 6 6 6
## [3049] 6 6 6 6 6 6 6 8 8 8 8 8 8 8 8 8 8 8 8 10 10 10 10
## [3073] 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 3
## [3097] 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3
## [3121] 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 7 7 7 7
## [3145] 7 7 7 7 7 7 7 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6
## [3169] 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 10 10 10 10 10 10 10
## [3193] 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 3 3 3 3
## [3217] 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 2 2
## [3241] 2 2 2 2 2 2 2 2 2 2 2 2 2 2 7 7 7 7 7 7 7 7
## [3265] 7 7 7 7 7 7 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6
## [3289] 8 8 8 8 8 8 8 8 8 8 8 8 8 8 10 10 10 10 10 10 10
## [3313] 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 3 3 3 3 3 3
## [3337] 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 2 2
## [3361] 2 2 2 2 2 2 2 2 2 2 2 7 7 7 7 7 7 7 7 7 7 7
## [3385] 7 7 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 8 8 8 8
## [3409] 8 8 8 8 8 8 8 8 8 8 10 10 10 10 10 10 10 10 10 10 10
## [3433] 10 10 10 10 10 10 10 10 10 10 10 10 10 10 3 3 3 3 3 3 3
## [3457] 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 2 2 2 2 2 2
## [3481] 2 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 6 6 6 6 6 6
## [3505] 6 6 6 6 6 6 6 6 6 6 8 8 8 8 8 8 8 8 8 8 8 8
## [3529] 8 8 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10
## [3553] 10 10 10 10 10 10 10 10 10 10 3 3 3 3 3 3 3 3 3 3 3
## [3577] 3 3 3 3 3 3 3 3 3 3 3 2 2 2 2 2 2 2 2 2 2 2
## [3601] 4 4 4 4 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 6 6
## [3625] 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 8 8 8 8 8 8 8
## [3649] 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 10 10 10 10 10 10
## [3673] 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 3
## [3697] 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 2
## [3721] 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2
## [3745] 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4
## [3769] 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4
## [3793] 4 4 4 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7
## [3817] 7 7 7 7 7 7 7 7 7 7 7 7 6 6 6 6 6 6 6 6 6 6
## [3841] 6 6 6 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8
## [3865] 8 8 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10 10
## [3889] 10 10 10 10 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3
## [3913] 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2
## [3937] 2 2 2 2 2 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4
## [3961] 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4
## [3985] 4 4 4 4 4 4 4 4 4 4 7 7 7 7 7 7 7 7 7 7 7
## [4009] 7 7 7 7 7 7 7 7 7 7 7 7 7 7

```

```

##
## Within cluster sum of squares by cluster:
## [1] 7663.348 32840.713 69902.707 34946.972 831375.777 6863.163
## [7] 22577.296 9307.018 308528.395 82293.150
## (between_SS / total_SS = 95.1 %)
##
## Available components:
##
## [1] "cluster" "centers" "totss" "withinss" "tot.withinss"
## [6] "betweenss" "size" "iter" "ifault"

```

```
cluster11 <- kmeans(datasetcluster_df[, 1:2],11 )
cluster11
```

```
## K-means clustering with 11 clusters of sizes 406, 553, 441, 307, 104, 467, 518, 341, 215, 209, 461
```

```
##
```

```
## Cluster means:
```

```
##           x           y
## 1  94.91133 143.7931
## 2  12.29295 143.4955
## 3  56.34467 223.0181
## 4  60.12052 139.9218
## 5 180.79808 134.5673
## 6 231.08994 224.3683
## 7 179.55019 225.2664
## 8  58.82698 186.9824
## 9 239.42326 139.0791
## 10 120.97129 141.4019
## 11 77.33189 149.5922
```

```
##
```

```
## Clustering vector:
```

```
## [1] 3 3 7 7 7 7 6 6 3 3 3 3 3 3 3 3 3 3 3 3 3
## [25] 3 3 3 3 3 3 3 7 7 7 7 7 7 7 7 7 7 7 7 7 7
## [49] 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7
## [73] 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 6 6 6 6 6
## [97] 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6
## [121] 6 6 6 6 6 6 6 6 6 6 6 6 3 7 7 7 7 7 7 7 7
## [145] 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7
## [169] 7 7 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6
## [193] 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 7
## [217] 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7
## [241] 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7
## [265] 7 7 7 7 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6
## [289] 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 3 3 3 3
## [313] 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 7 7 7 7 7
## [337] 7 7 7 7 7 7 7 7 7 7 7 7 7 7 6 6 6 6 6 6 6
## [361] 6 6 6 6 6 6 6 6 6 6 6 6 6 3 3 3 3 3 3 3 3
## [385] 3 3 3 3 3 3 3 3 3 3 7 7 7 7 7 7 7 7 7 7 7
## [409] 7 7 7 7 7 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6
## [433] 6 6 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3
## [457] 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 6 6 6 6
## [481] 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 3 3 3 3 3 3
## [505] 3 3 3 3 3 3 7 7 7 7 7 7 7 7 7 7 7 6 6 6 6
## [529] 6 6 6 6 6 6 6 6 6 6 6 3 3 3 3 3 3 3 3 3 7
## [553] 7 7 7 7 7 7 7 7 7 7 7 6 6 6 6 6 6 6 6 6 3
## [577] 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 7 7 7 7 7 7
## [601] 7 7 7 7 7 6 6 6 6 6 6 6 6 6 6 6 6 6 6 3 3
## [625] 3 3 3 3 3 3 3 3 3 3 3 3 7 7 7 7 7 7 7 7 7
## [649] 7 7 6 6 6 6 6 6 6 6 6 6 6 6 6 3 3 3 3 3 3
## [673] 3 3 3 3 3 3 7 7 7 7 7 7 7 7 7 7 7 6 6 6 6
## [697] 6 6 6 6 6 6 6 6 6 6 6 3 3 3 3 3 3 3 3 3 3
## [721] 3 3 7 7 7 7 7 7 7 7 7 7 7 7 7 6 6 6 6 6 6
## [745] 6 6 6 6 6 6 6 6 3 3 3 3 3 3 3 3 3 3 3 3 3
## [769] 7 7 7 7 7 7 7 7 7 7 7 7 7 6 6 6 6 6 6 6 6
```

```

## [793] 6 6 6 6 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 7 7 7 7
## [817] 7 7 7 7 7 7 7 7 7 7 6 6 6 6 6 6 6 6 6 6 6 6
## [841] 6 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 7 7 7 7 7 7
## [865] 7 7 7 7 7 7 7 7 6 6 6 6 6 6 6 6 6 6 6 6 6 3
## [889] 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 7 7 7 7 7 7
## [913] 7 7 7 7 7 7 7 7 7 7 7 6 6 6 6 6 6 6 6 6 6 6
## [937] 6 6 6 6 6 6 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3
## [961] 3 3 3 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 6
## [985] 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 3 3 3 3
## [1009] 3 3 3 3 3 3 3 3 3 3 3 3 3 3 7 7 7 7 7 7 7 7
## [1033] 7 7 7 7 7 7 7 7 7 7 6 6 6 6 6 6 6 6 6 6 6 6
## [1057] 6 6 6 6 6 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3
## [1081] 3 3 3 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 6 6
## [1105] 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 3 3 3 3
## [1129] 3 3 3 3 3 3 3 3 3 3 3 3 3 3 7 7 7 7 7 7 7 7
## [1153] 7 7 7 7 7 7 7 7 7 7 6 6 6 6 6 6 6 6 6 6 6 6
## [1177] 6 6 6 6 6 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3
## [1201] 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 6 6 6 6
## [1225] 6 6 6 6 6 6 6 6 6 6 6 6 6 3 3 3 3 3 3 3 3 3
## [1249] 3 3 3 3 3 3 3 3 7 7 7 7 7 7 7 7 7 7 7 7 7 7
## [1273] 7 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 3 3 3 3
## [1297] 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 7 7 7 7 7 7
## [1321] 7 7 7 7 7 7 7 7 7 7 7 7 7 7 6 6 6 6 6 6 6 6
## [1345] 6 6 6 6 6 6 6 6 6 6 6 6 6 3 3 3 3 3 3 3 3 3
## [1369] 3 3 3 3 3 3 3 3 3 3 3 3 7 7 7 7 7 7 7 7 7 7
## [1393] 7 7 7 7 7 7 7 7 7 7 7 6 6 6 6 6 6 6 6 6 6 6
## [1417] 6 6 6 6 6 6 6 6 6 6 3 8 8 8 8 8 8 8 8 8 8 8
## [1441] 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8
## [1465] 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8
## [1489] 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8
## [1513] 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8
## [1537] 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8
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## [1585] 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8
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## [4009]  9  9  9  9  9  9  9  9  9  9  9  9  9  9  9  9  9  9  9  9  9  9  9  9
##
## Within cluster sum of squares by cluster:
## [1] 32021.43 62364.78 47345.46 24864.66 32104.29 65846.87 112959.43
## [8] 35774.69 21706.14 29808.07 39943.55
## (between_SS / total_SS =  98.2 %)
##
## Available components:
##
## [1] "cluster"      "centers"      "totss"        "withinss"     "tot.withinss"
## [6] "betweenss"    "size"         "iter"         "ifault"

```

```

cluster12 <- kmeans(datasetcluster_df[, 1:2],12 )
cluster12

```

```

## K-means clustering with 12 clusters of sizes 105, 521, 341, 95, 233, 107, 101, 218, 985, 140, 616, 5
##
## Cluster means:
##           x           y
## 1  63.12381 229.8000
## 2  65.72169 143.4299
## 3  58.82698 186.9824
## 4  64.22105 216.7684
## 5 118.17167 141.4549
## 6 178.21495 134.5514
## 7  49.79208 229.6634
## 8 239.00917 139.0229
## 9 203.98579 224.8406

```

```
## 10 50.64286 217.3786
## 11 89.27110 146.7792
## 12 12.61071 143.3839
```

```
##
```

```
## Clustering vector:
```

```
## [1] 7 1 9 9 9 9 9 9 7 7 7 7 7 7 7 7 7 1 1 1 1 1
## [25] 1 1 1 1 1 1 1 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9
## [49] 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9
## [73] 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9
## [97] 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9
## [121] 9 9 9 9 9 9 9 9 9 9 9 9 7 9 9 9 9 9 9 9 9 9
## [145] 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9
## [169] 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9
## [193] 7 7 7 7 7 7 7 7 7 7 7 1 1 1 1 1 1 1 1 1 1 9
## [217] 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9
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## [3433] 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 11 11 11 11 11 11
## [3457] 11 11 11 11 11 11 11 11 11 11 11 11 11 11 5 5 5 5 5 5 5
## [3481] 5 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 12 12 12 12 12 12
## [3505] 12 12 12 12 12 12 12 12 12 12 12 12 12 12 12 12 12 12 12 12
## [3529] 12 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2
## [3553] 2 2 2 2 2 2 2 2 2 2 2 11 11 11 11 11 11 11 11 11 11
## [3577] 11 11 11 11 11 11 11 11 11 11 5 5 5 5 5 5 5 5 5 5 5
## [3601] 6 6 6 6 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 12 12
## [3625] 12 12 12 12 12 12 12 12 12 12 12 12 12 12 12 12 12 12 12 12
## [3649] 12 12 12 12 12 12 12 12 12 12 2 2 2 2 2 2 2 2 2 2 2
## [3673] 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 11
## [3697] 11 11 11 11 11 11 11 11 11 11 11 11 11 11 11 11 11 11 11 5
## [3721] 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 6
## [3745] 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6

```

```

## [3769] 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6
## [3793] 6 6 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8
## [3817] 8 8 8 8 8 8 8 8 8 8 8 8 12 12 12 12 12 12 12 12 12 12
## [3841] 12 12 12 12 12 12 12 12 12 12 12 12 12 12 12 12 12 12 12 12 2 2
## [3865] 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2
## [3889] 2 2 2 2 2 11 11 11 11 11 11 11 11 11 11 11 11 11 11 11 11 5
## [3913] 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5 5
## [3937] 5 5 5 5 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6
## [3961] 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6
## [3985] 6 6 6 6 6 6 6 6 8 8 8 8 8 8 8 8 8 8 8 8 8 8
## [4009] 8 8 8 8 8 8 8 8 8 8 8 8 8 8 8
##
## Within cluster sum of squares by cluster:
## [1] 2502.190 60236.338 35774.686 1735.263 32322.910 35016.523
## [7] 2441.188 24436.867 831375.777 3607.079 73385.700 67391.591
## (between_SS / total_SS = 95.9 %)
##
## Available components:
##
## [1] "cluster" "centers" "totss" "withinss" "tot.withinss"
## [6] "betweenss" "size" "iter" "ifault"

```

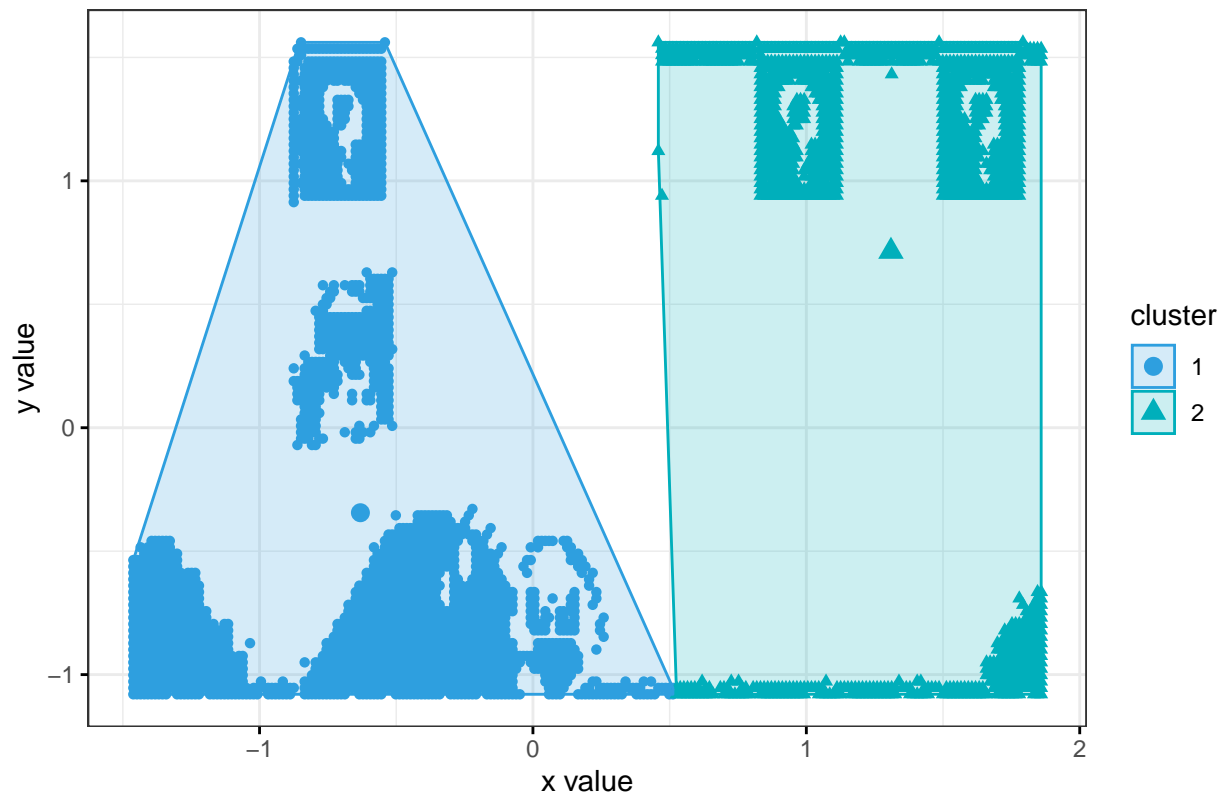
```
## fviz_cluster
```

```

fviz_cluster(cluster2, data= datasetcluster_df[, -100],
  palette = c("#2E9FDF", "#00AFBB", "#E7B800"),
  geom = "point",
  ellipse.type = "convex",
  ggtheme = theme_bw())

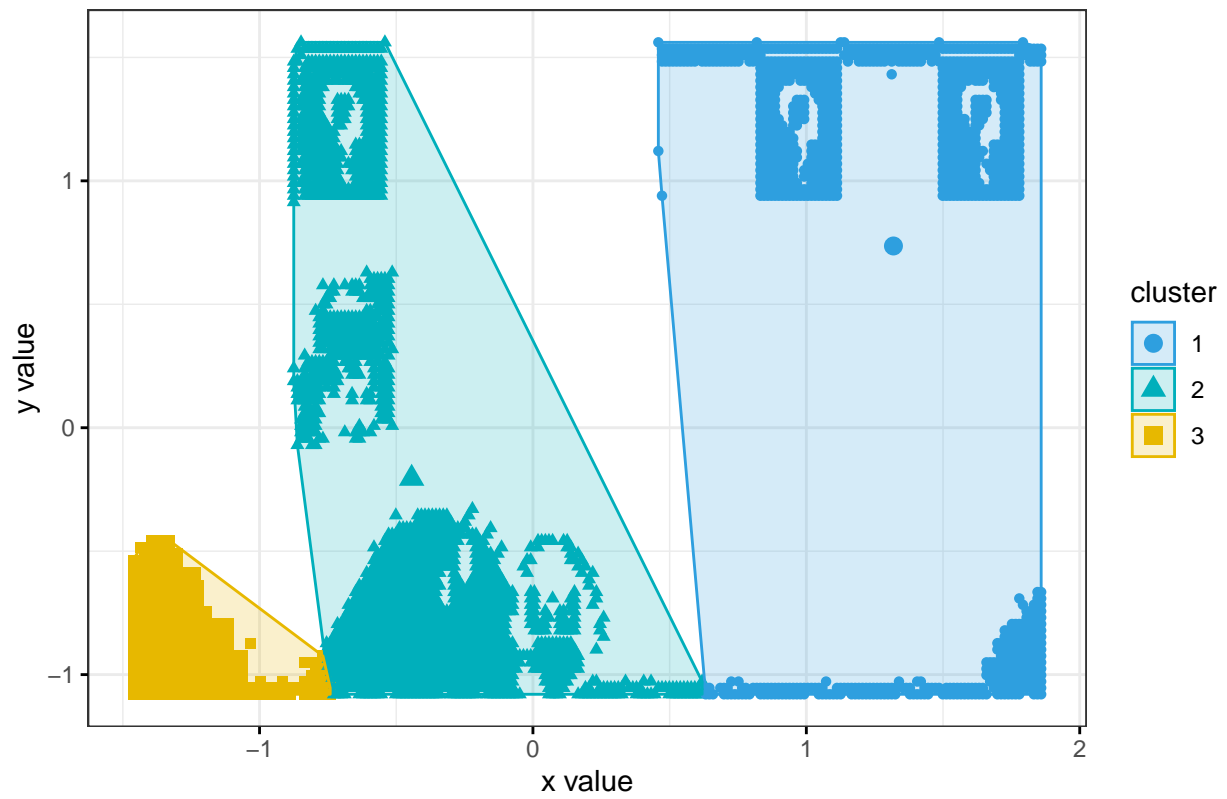
```

Cluster plot



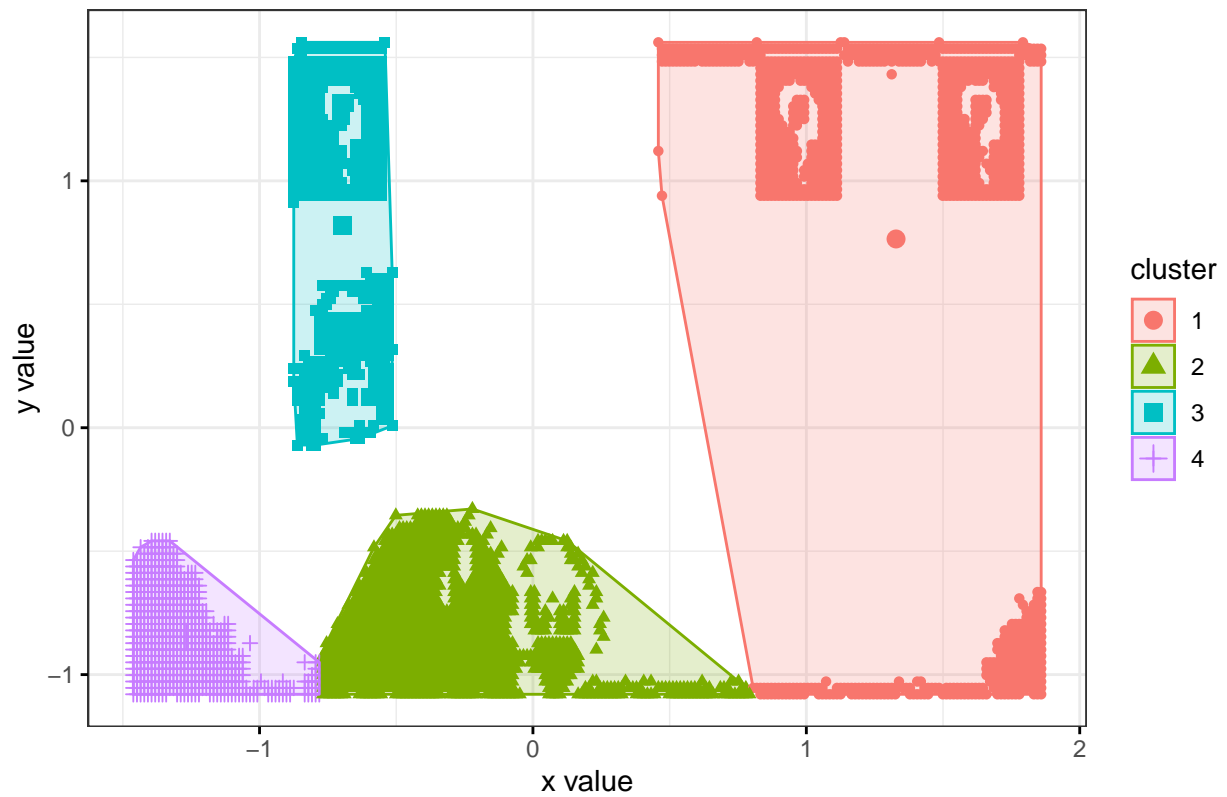
```
fviz_cluster(cluster3, data= datasetcluster_df[, -100],
  palette = c("#2E9FDF", "#00AFBB", "#E7B800"),
  geom = "point",
  ellipse.type = "convex",
  ggtheme = theme_bw())
```

Cluster plot



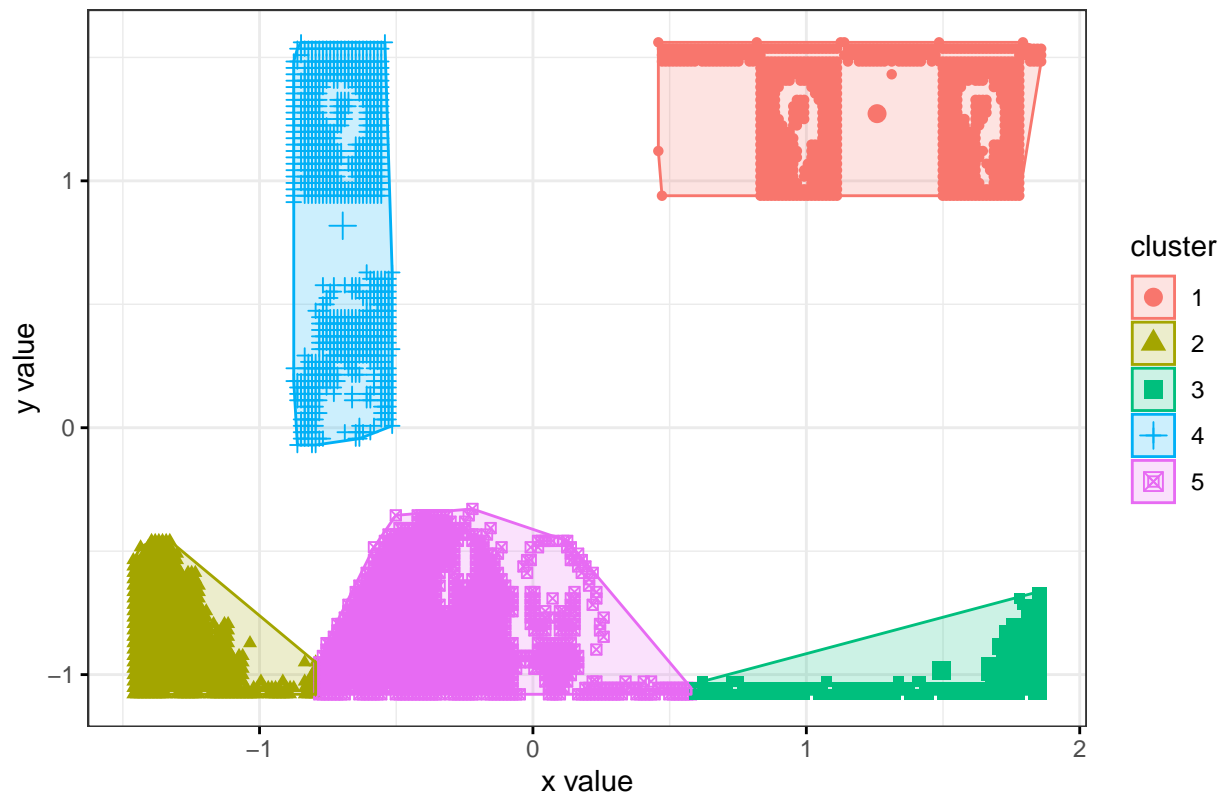
```
fviz_cluster(cluster4, data= datasetcluster_df[, -100],  
             geom = "point",  
             ellipse.type = "convex",  
             ggtheme = theme_bw())
```

Cluster plot



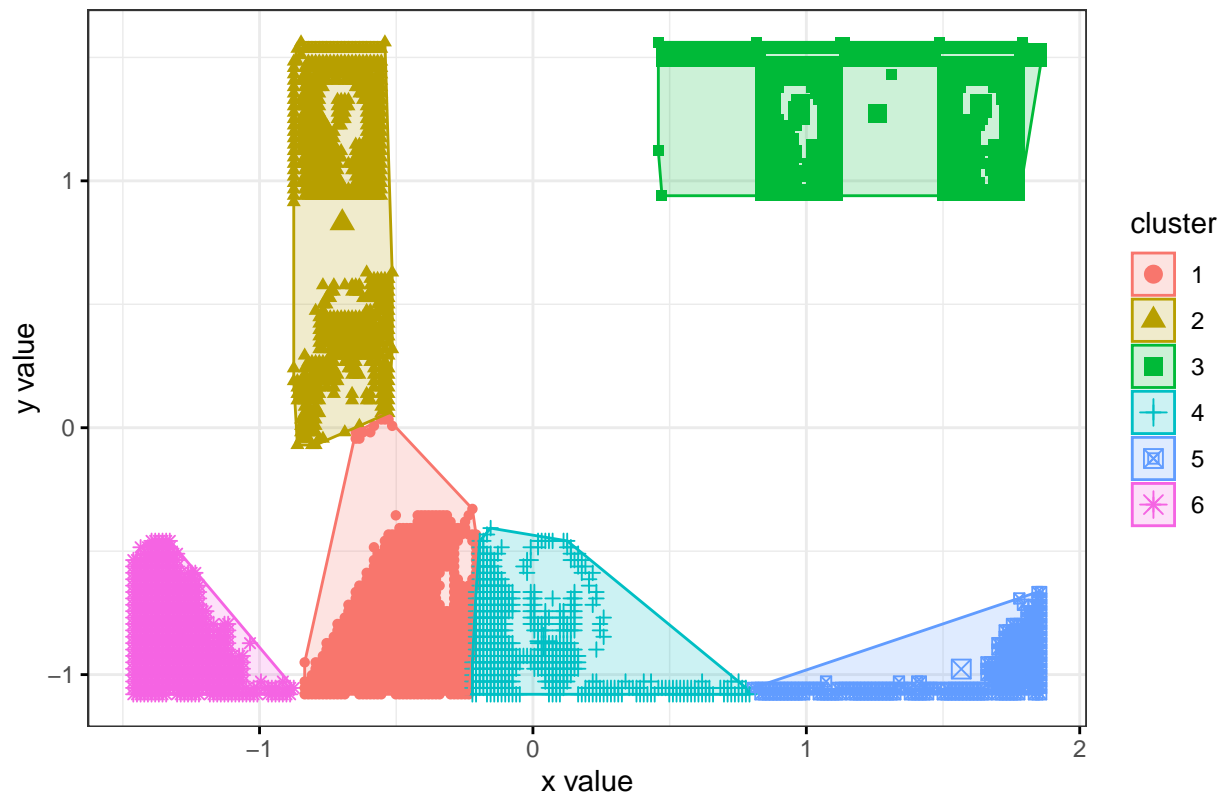
```
fviz_cluster(cluster5, data= datasetcluster_df[, -100],  
             geom = "point",  
             ellipse.type = "convex",  
             ggtheme = theme_bw())
```

Cluster plot



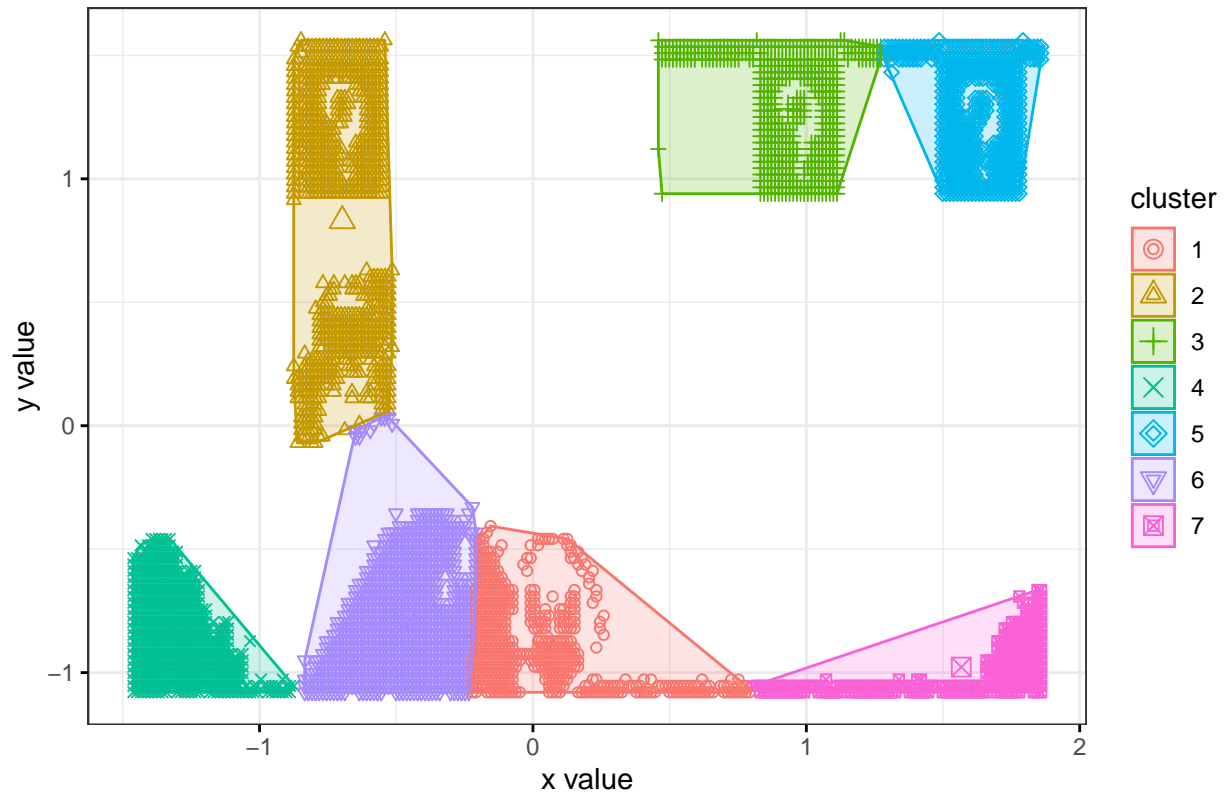
```
fviz_cluster(cluster6, data= datasetcluster_df[, -100],
             geom = "point",
             ellipse.type = "convex",
             ggtheme = theme_bw())
```

Cluster plot



```
fviz_cluster(cluster7, data= datasetcluster_df[, -100],  
             geom = "point",  
             ellipse.type = "convex",  
             ggtheme = theme_bw())
```

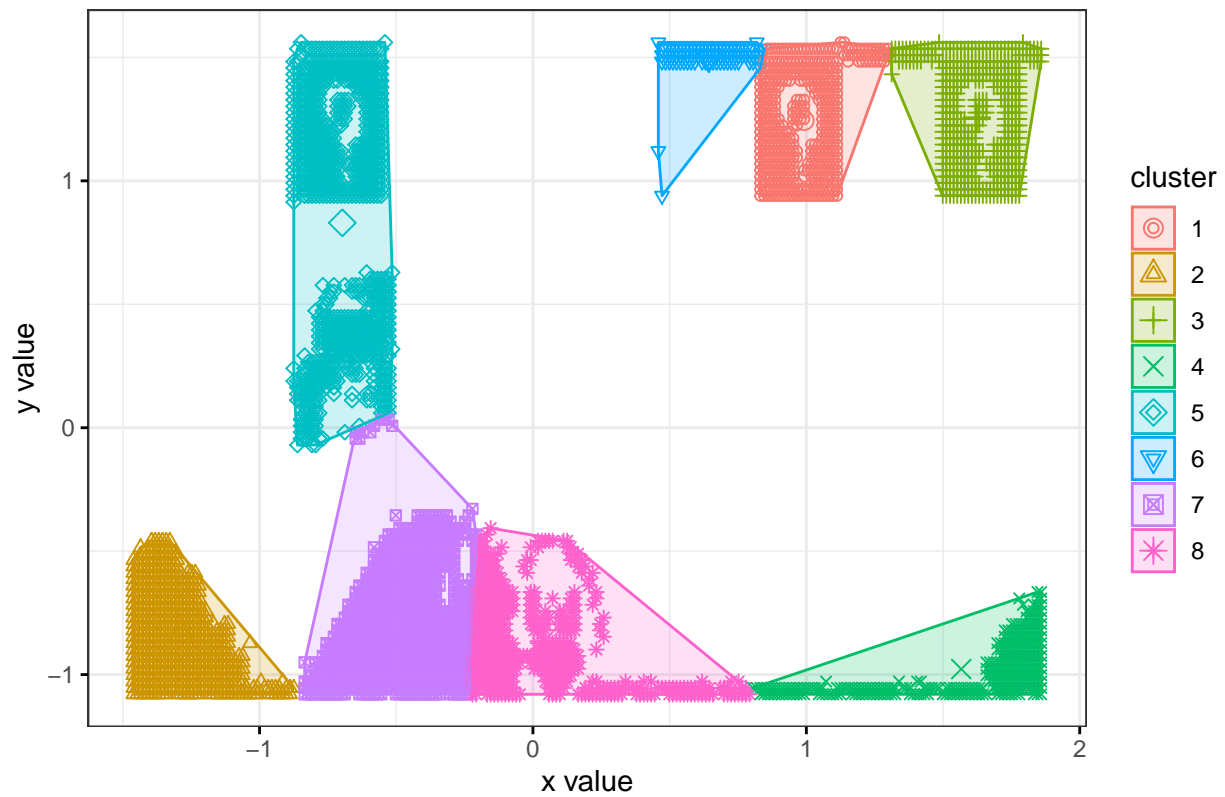
Cluster plot



```
fviz_cluster(cluster8, data= datasetcluster_df[, -100],
             geom = "point",
             ellipse.type = "convex",
             ggtheme = theme_bw())
```

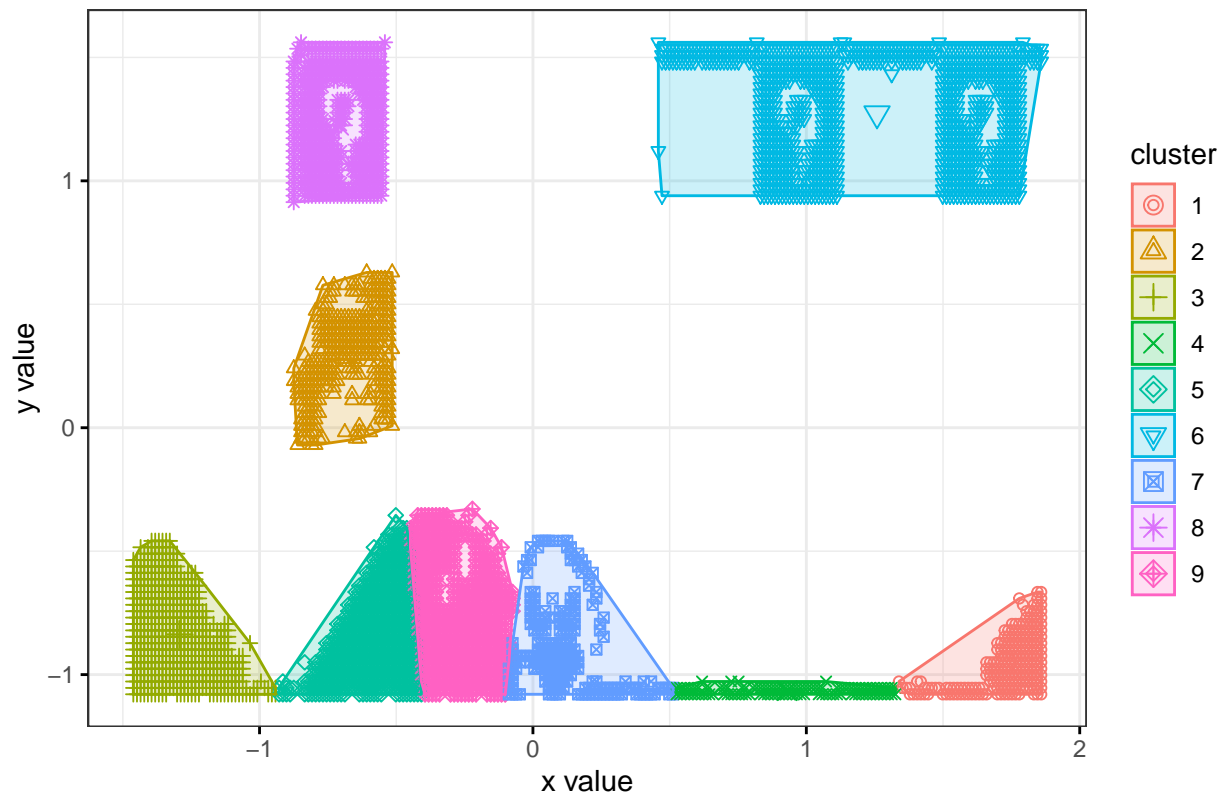


Cluster plot



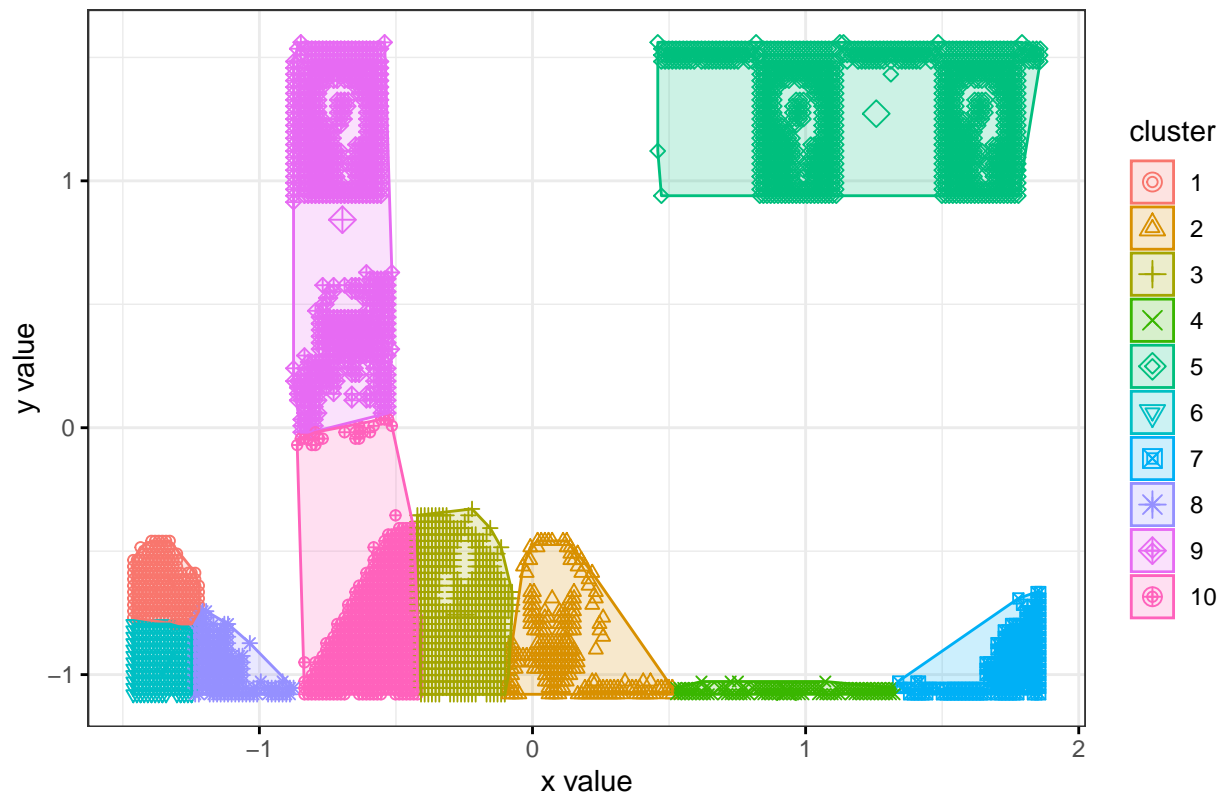
```
fviz_cluster(cluster9, data= datasetcluster_df[, -100],
  geom = "point",
  ellipse.type = "convex",
  ggtheme = theme_bw())
```

Cluster plot



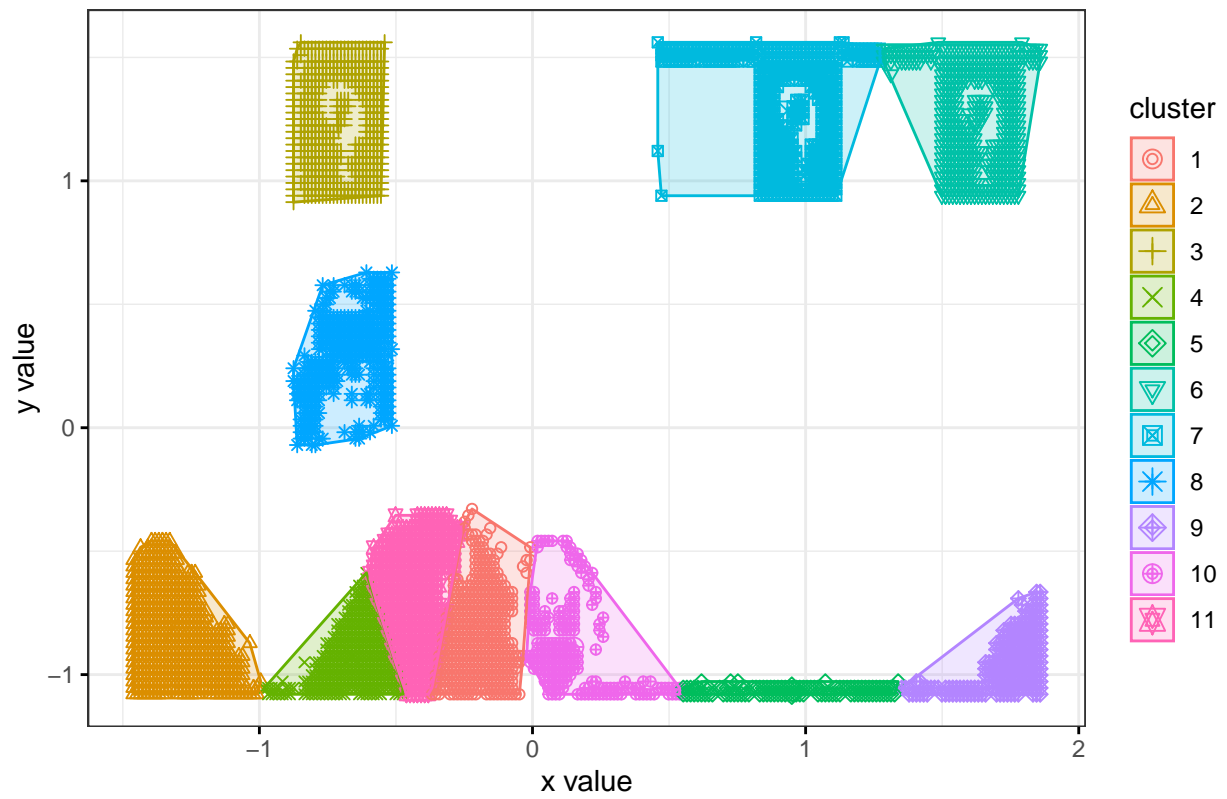
```
fviz_cluster(cluster10, data= datasetcluster_df[, -100],
  geom = "point",
  ellipse.type = "convex",
  ggtheme = theme_bw())
```

Cluster plot

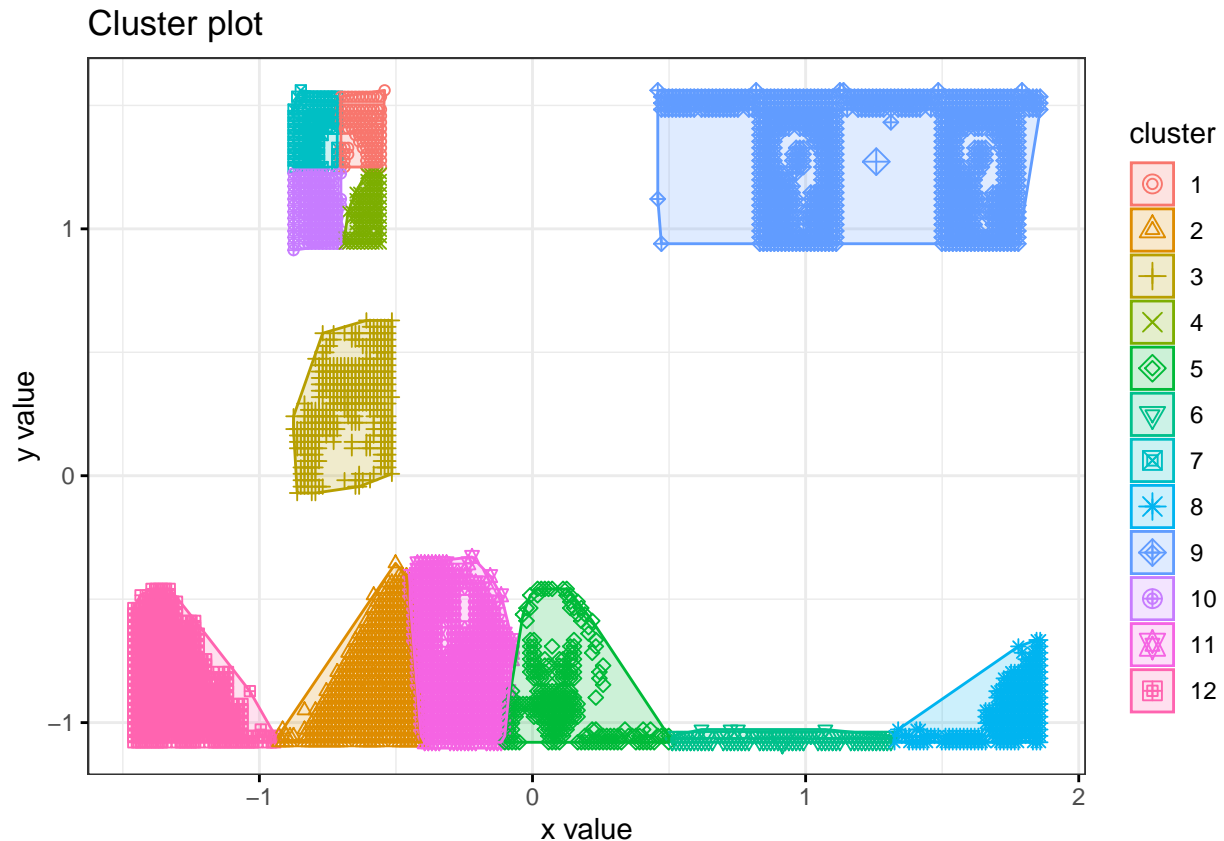


```
fviz_cluster(cluster11, data= datasetcluster_df[, -100],
  geom = "point",
  ellipse.type = "convex",
  ggtheme = theme_bw())
```

Cluster plot



```
fviz_cluster(cluster12, data= datasetcluster_df[, -100],
             geom = "point",
             ellipse.type = "convex",
             ggtheme = theme_bw())
```



C. As k-means is an unsupervised algorithm, you cannot compute the accuracy as there are no correct values to compare the output to. Instead,

you will use the average distance from the center of each cluster as a measure of how well the model fits the data. To calculate this metric,

simply compute the distance of each data point to the center of the cluster it is assigned to and take the average value of all of those distances.

```
distance2<-distance(cluster2$centers, method = "euclidean")
```

```
## Metric: 'euclidean'; comparing: 2 vectors.
```

```
distance2
```

```
## euclidean  
## 151.1523
```

```
distance3<-distance(cluster3$centers, method = "euclidean")
```

```
## Metric: 'euclidean'; comparing: 3 vectors.
```

```
distance3
```

```
##          v1          v2          v3  
## v1  0.0000 137.06041 202.68484  
## v2 137.0604  0.00000  65.94272  
## v3 202.6848  65.94272  0.00000
```

```
distance4<-distance(cluster4$centers, method = "euclidean")
```

```
## Metric: 'euclidean'; comparing: 4 vectors.
```

```
distance4
```

```
##          v1          v2          v3          v4  
## v1  0.0000 135.44096 151.78432 204.43178  
## v2 135.4410  0.00000  69.87657  73.65898  
## v3 151.7843  69.87657  0.00000  77.35550  
## v4 204.4318  73.65898  77.35550  0.00000
```

```
distance5<-distance(cluster5$centers, method = "euclidean")
```

```
## Metric: 'euclidean'; comparing: 5 vectors.
```

```
distance5
```

```
##          v1          v2          v3          v4          v5  
## v1  0.00000 206.75345  88.93857 147.60409 142.17903  
## v2 206.75345  0.00000 207.58334  77.50881  72.46763  
## v3  88.93857 207.58334  0.00000 178.34938 135.24745  
## v4 147.60409  77.50881 178.34938  0.00000  69.06147  
## v5 142.17903  72.46763 135.24745  69.06147  0.00000
```

```
distance6<-distance(cluster6$centers, method = "euclidean")
```

```
## Metric: 'euclidean'; comparing: 6 vectors.
```

```
distance6
```

```
##          v1          v2          v3          v4          v5          v6
## v1  0.00000  64.20602 151.86600  38.41595 152.99891  61.27086
## v2  64.20602   0.00000 147.66212  85.63776 183.56123  78.22516
## v3 151.86600 147.66212   0.00000 123.18258  89.91041 207.61918
## v4  38.41595  85.63776 123.18258   0.00000 114.62485  99.47405
## v5 152.99891 183.56123  89.91041 114.62485   0.00000 214.07714
## v6  61.27086  78.22516 207.61918  99.47405 214.07714   0.00000
```

```
distance7<-distance(cluster7$centers, method = "euclidean")
```

```
## Metric: 'euclidean'; comparing: 7 vectors.
```

```
distance7
```

```
##          v1          v2          v3          v4          v5          v6          v7
## v1  0.00000  85.63776 106.64207  99.47405 144.18000  38.41595 114.62485
## v2  85.63776   0.00000 123.48264  78.22516 174.56686  64.20602 183.56123
## v3 106.64207 123.48264   0.00000 185.58640  51.54757 131.86874  99.41820
## v4  99.47405  78.22516 185.58640   0.00000 232.62119  61.27086 214.07714
## v5 144.18000 174.56686  51.54757 232.62119   0.00000 175.36875  86.51339
## v6  38.41595  64.20602 131.86874  61.27086 175.36875   0.00000 152.99891
## v7 114.62485 183.56123  99.41820 214.07714  86.51339 152.99891   0.00000
```

```
distance8<-distance(cluster8$centers, method = "euclidean")
```

```
## Metric: 'euclidean'; comparing: 8 vectors.
```

```
distance8
```

```
##          v1          v2          v3          v4          v5          v6          v7
## v1  0.00000 188.95086  47.35289  96.02431 127.69927  27.94801 134.57937
## v2 188.95086   0.00000 232.83459 214.07714  78.22516 170.68144  61.27086
## v3  47.35289 232.83459   0.00000  86.41990 174.82278  74.09473 175.56015
## v4  96.02431 214.07714  86.41990   0.00000 183.56123 118.09989 152.99891
## v5 127.69927  78.22516 174.82278 183.56123   0.00000 103.81775  64.20602
## v6  27.94801 170.68144  74.09473 118.09989 103.81775   0.00000 121.20981
## v7 134.57937  61.27086 175.56015 152.99891  64.20602 121.20981   0.00000
## v8 108.36619  99.47405 144.33935 114.62485  85.63776 101.94947  38.41595
##          v8
## v1 108.36619
## v2  99.47405
## v3 144.33935
## v4 114.62485
## v5  85.63776
## v6 101.94947
## v7  38.41595
## v8   0.00000
```

```
distance9<-distance(cluster9$centers, method = "euclidean")
```

```
## Metric: 'euclidean'; comparing: 9 vectors.
```

```
distance9
```

```
##          v1          v2          v3          v4          v5          v6          v7
## v1  0.00000 186.71349 226.71747  60.10161 173.53092  92.75594 120.68118
## v2 186.71349  0.00000  63.53558 131.43838  44.09338 150.01441  75.16694
## v3 226.71747  63.53558  0.00000 166.97819  53.20075 207.98944 106.03634
## v4  60.10161 131.43838 166.97819  0.00000 113.89154  93.58851  61.11574
## v5 173.53092  44.09338  53.20075 113.89154  0.00000 160.36609  52.85580
## v6  92.75594 150.01441 207.98944  93.58851 160.36609  0.00000 119.33452
## v7 120.68118  75.16694 106.03634  61.11574  52.85580 119.33452  0.00000
## v8 201.28604  36.12113  90.85300 151.51878  80.13358 147.65236 102.63392
## v9 150.07058  50.54412  76.87572  90.76466  23.83080 138.65837  29.69298
##          v8          v9
## v1 201.28604 150.07058
## v2  36.12113  50.54412
## v3  90.85300  76.87572
## v4 151.51878  90.76466
## v5  80.13358  23.83080
## v6 147.65236 138.65837
## v7 102.63392  29.69298
## v8  0.00000  83.13482
## v9  83.13482  0.00000
```

```
distance10<-distance(cluster10$centers, method = "euclidean")
```

```
## Metric: 'euclidean'; comparing: 10 vectors.
```

```
distance10
```

```
##          v1          v2          v3          v4          v5          v6          v7
## v1  0.00000 110.78876  81.56189 171.75654 209.14323  11.46448 231.21443
## v2 110.78876  0.00000  29.37351  60.98459 119.21337 110.97879 120.54598
## v3  81.56189  29.37351  0.00000  90.32648 138.48693  82.31463 149.65304
## v4 171.75654  60.98459  90.32648  0.00000  93.58851 171.61718  60.10161
## v5 209.14323 119.21337 138.48693  93.58851  0.00000 213.98283  92.75594
## v6  11.46448 110.97879  82.31463 171.61718 213.98283  0.00000 231.48114
## v7 231.21443 120.54598 149.65304  60.10161  92.75594 231.48114  0.00000
## v8  20.55292  93.95859  65.44323 154.55496 198.83130  17.06622 214.43557
## v9  75.49987  90.67613  69.77051 142.48227 147.51180  84.86390 194.58912
## v10 58.12769  52.67048  23.62261 113.62902 159.11841  58.71021 173.16905
##          v8          v9          v10
## v1  20.55292  75.49987  58.12769
## v2  93.95859  90.67613  52.67048
## v3  65.44323  69.77051  23.62261
## v4 154.55496 142.48227 113.62902
## v5 198.83130 147.51180 159.11841
## v6  17.06622  84.86390  58.71021
## v7 214.43557 194.58912 173.16905
## v8  0.00000  76.91907  41.90640
## v9  76.91907  0.00000  63.64930
## v10 41.90640  63.64930  0.00000
```



```
distance11<- distance(cluster11$centers, method = "euclidean")
```

```
## Metric: 'euclidean'; comparing: 11 vectors.
```

```
distance11
```

```
##          v1          v2          v3          v4          v5          v6          v7
## v1    0.00000  82.61892  88.11353  35.00553  86.38083 158.23077 117.48037
## v2    82.61892   0.00000  90.90879  47.96090 168.74149 233.26495 186.17591
## v3    88.11353  90.90879   0.00000  83.18206 152.68333 174.75048 123.22603
## v4    35.00553  47.96090  83.18206   0.00000 120.79629 190.68757 146.78946
## v5    86.38083 168.74149 152.68333 120.79629   0.00000 102.92469  90.70769
## v6   158.23077 233.26495 174.75048 190.68757 102.92469   0.00000  51.54757
## v7   117.48037 186.17591 123.22603 146.78946  90.70769  51.54757   0.00000
## v8    56.27962  63.69088  36.12113  47.07835 132.75651 176.27317 126.64817
## v9   144.58879 227.17324 201.40391 179.30472  58.79853  85.69538 104.94304
## v10   26.16944 108.69851 104.10480  60.86877  60.21591 137.87508 102.29732
## v11   18.51125  65.32407  76.36644  19.74201 104.55142 170.97662 127.18164
##          v8          v9          v10          v11
## v1    56.27962 144.58879  26.16944  18.51125
## v2    63.69088 227.17324 108.69851  65.32407
## v3    36.12113 201.40391 104.10480  76.36644
## v4    47.07835 179.30472  60.86877  19.74201
## v5   132.75651  58.79853  60.21591 104.55142
## v6   176.27317  85.69538 137.87508 170.97662
## v7   126.64817 104.94304 102.29732 127.18164
## v8     0.00000 186.84150  77.06813  41.71882
## v9   186.84150   0.00000 118.47474 162.43195
## v10   77.06813 118.47474   0.00000  44.40133
## v11   41.71882 162.43195  44.40133   0.00000
```

```
distance12<- distance(cluster12$centers, method = "euclidean")
```

```
## Metric: 'euclidean'; comparing: 12 vectors.
```

```
distance12
```

```
##          v1          v2          v3          v4          v5          v6          v7
## v1    0.00000  86.40912  43.03265  13.07769 104.09187 149.39299  13.33243
## v2    86.40912   0.00000  44.09483  73.35383  52.48716 112.84309  87.69239
## v3    43.03265  44.09483   0.00000  30.27049  74.79668 130.39363  43.62676
## v4    13.07769  73.35383  30.27049   0.00000  92.64335 140.54980  19.35135
## v5   104.09187  52.48716  74.79668  92.64335   0.00000  60.43885 111.60867
## v6   149.39299 112.84309 130.39363 140.54980  60.43885   0.00000 159.80839
## v7    13.33243  87.69239  43.62676  19.35135 111.60867 159.80839   0.00000
## v8   197.92963 173.34352 186.45572 191.29884 120.86197  60.95844 209.80657
## v9   140.94925 160.45142 150.01441 139.99765 119.65464  93.89503 154.26911
## v10   17.60869  75.47033  31.47867  13.59190 101.60974 152.10187  12.31422
## v11   87.04097  23.78640  50.42956  74.33702  29.38692  89.78044  91.80618
## v12  100.09650  53.11099  63.53558  89.71572 105.57858 165.83961  93.94996
##          v8          v9          v10          v11          v12
```

```
## v1 197.92963 140.94925 17.60869 87.04097 100.09650
## v2 173.34352 160.45142 75.47033 23.78640 53.11099
## v3 186.45572 150.01441 31.47867 50.42956 63.53558
## v4 191.29884 139.99765 13.59190 74.33702 89.71572
## v5 120.86197 119.65464 101.60974 29.38692 105.57858
## v6 60.95844 93.89503 152.10187 89.78044 165.83961
## v7 209.80657 154.26911 12.31422 91.80618 93.94996
## v8 0.00000 92.68932 204.01342 149.93882 226.44046
## v9 92.68932 0.00000 153.52438 138.75532 207.98944
## v10 204.01342 153.52438 0.00000 80.47614 83.19646
## v11 149.93882 138.75532 80.47614 0.00000 76.73554
## v12 226.44046 207.98944 83.19646 76.73554 0.00000
```

```
mean2<- mean(distance2)
mean2
```

```
## [1] 151.1523
```

```
mean3<- mean(distance3)
mean3
```

```
## [1] 90.15288
```

```
mean4<- mean(distance4)
mean4
```

```
## [1] 89.06851
```

```
mean5<- mean(distance5)
mean5
```

```
## [1] 106.0555
```

```
mean6<- mean(distance6)
mean6
```

```
## [1] 100.7073
```

```
mean7<- mean(distance7)
mean7
```

```
## [1] 106.2975
```

```
mean8<- mean(distance8)
mean8
```

```
## [1] 103.9763
```

```
mean9<- mean(distance9)
mean9
```

```
## [1] 96.63894
```

```
mean10<- mean(distance10)
mean10
```

```
## [1] 99.46185
```

```
mean11<- mean(distance11)
mean11
```

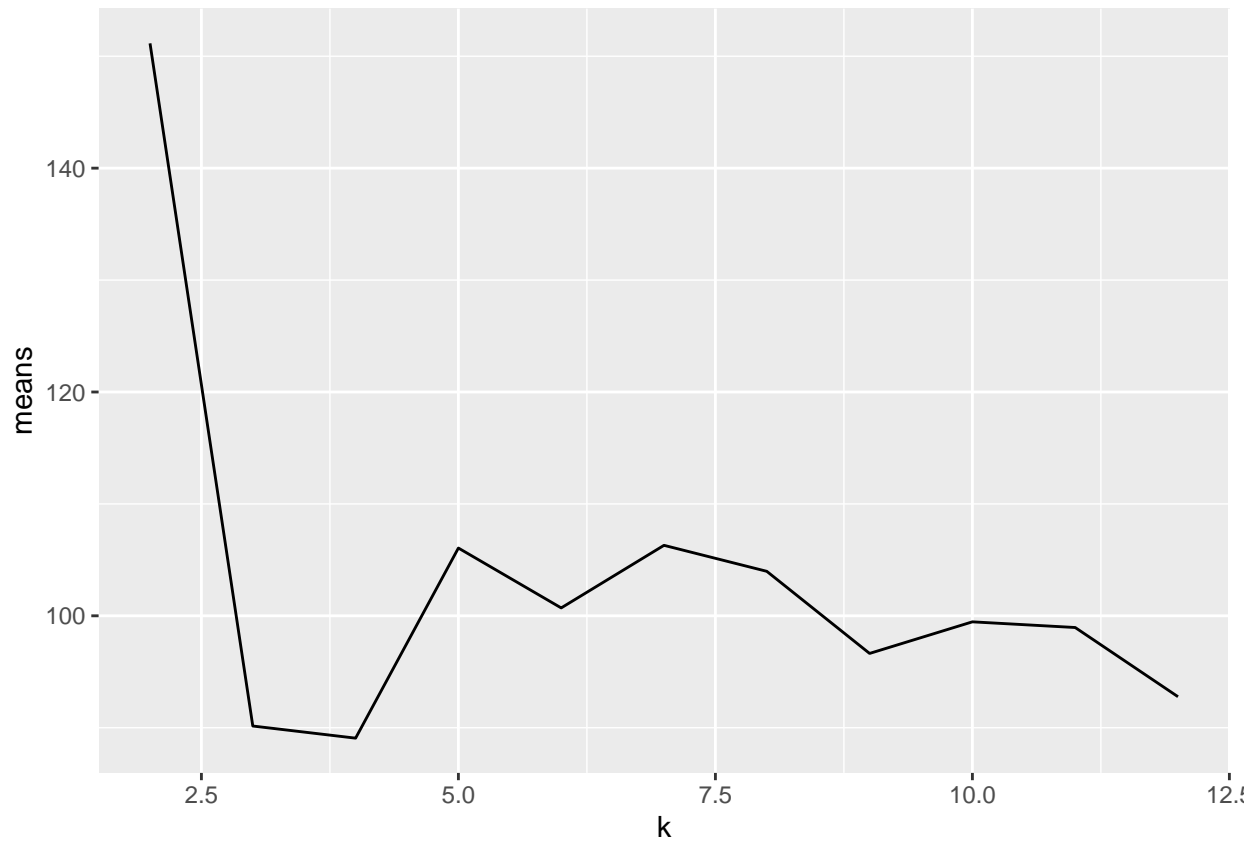
```
## [1] 98.95458
```

```
mean12<- mean(distance12)
mean12
```

```
## [1] 92.77099
```

```
averagedf<- data.frame('k' = 2:12,
                       'means'= c(mean2,
                                   mean3,
                                   mean4,
                                   mean5,
                                   mean6,
                                   mean7,
                                   mean8,
                                   mean9,
                                   mean10,
                                   mean11,
                                   mean12))

ggplot(averagedf,aes(x=k, y=means)) +geom_line()
```



averagedf

```
##      k      means
## 1     2 151.15225
## 2     3  90.15288
## 3     4  89.06851
## 4     5 106.05546
## 5     6 100.70735
## 6     7 106.29746
## 7     8 103.97633
## 8     9  96.63894
## 9    10  99.46185
## 10   11  98.95458
## 11   12  92.77099
```

Looking at the graph creating the elbow point would be where  $k=3$  and the average distance is around 88.69. This is where the graph changes shape and has an inflection point.